```
function est=imp_sampling(param, param_prop, data, D, M, ub, 1b, seed, options)
    consumer = data(:,1);
   N_cons = length(unique(consumer));
   N prod = data(:,end-2);
   Js = unique(N prod);
   Num J = length(Js);
   table=importdata('tableZ.csv');%comment unless using look-up table method
   W_PROP = zeros(N_cons, M, Num_J);
    L_ALL = zeros(N_cons, M, Num_J);
   C = \{\};
   WEIGHT = \{\};
   % Proposal paramters
   theta mean1 = param prop(1, 1:4);
   theta std1 = param prop(1, 6:9);
   c_mean1 = param_prop(1, 5);
    c_{std1} = param_prop(1, 10);
   %construct likelihood for consumers with the same number of searches
    for i = 1:Num J
        nalt = Js(i);
        dat = data(N prod == nalt,:);
        N_obs=length(dat);
        % Choices
        tran = dat(:, end);
        searched = dat(:, end - 1);
        last = dat(:, end - 4);
        has_searched = dat(:, end - 3);
        outside = dat(:, 3);
        X = dat(:,4:7);
        % Simulate Proposal draws
        theta all = zeros(N obs, M);
        c_all = zeros(N_obs, M);
        theta weight = zeros(N cons*(nalt - 1), M);
        for pd = 1:M
            rng(pd*seed, 'twister');
            b = theta mean1 + theta std1.*randn(N cons, nalt - 1);
            theta draws = X.*repelem(b, nalt, 1);
            c_draws = exp(c_mean1 + c_std1*randn(N_obs, 1));
            theta all(:, pd) = sum(theta draws,2);
            theta_weight(:, pd) = reshape(b', N_cons*(nalt - 1), 1);
            c_all(:, pd) = c_draws;
        end
        C\{i\} = c all;
        WEIGHT{i} = theta weight;
        % Compute poposal weights
```

```
for pd = 1:M
    theta_pd = reshape(theta_weight(:, pd), nalt - 1, N_cons);
    for cons = 1:N_cons
        theta n pd = theta pd(:, cons)';
        c n pd = c all(consumer == cons, pd);
       w_theta = normpdf(theta_n_pd, theta_mean1, theta_std1);
        w_c = normpdf(log(c_n_pd), c_mean1, c_std1);
        W_PROP(cons, pd, i) = prod(w_theta)*prod(w_c);
    end
end
% Compute likelihood at proposal draws
for pd = 1:M
    rng(seed*pd);
    epsilonDraw=randn(N_obs,D); %D draws for every M
    etaDraw=randn(N_obs,D);
    c = c_all(:, pd);
    eut = (repmat(theta_all(:, pd), 1, D) + etaDraw).*(1 - outside);
    ut = eut + epsilonDraw;
    %%%1. look-up table method
    m=zeros(N_obs,1);
    for j=1:N_obs
        lookupvalue=abs(table(:,2)-c(j));
        if (table(1,2)>=c(j)\&\&c(j)>=table(end,2))
            [~,index_m]=min(lookupvalue);
            m(j)=table(index m,1);
        elseif table(1,2)<c(j)</pre>
            m(j)=-c(j);
        elseif c(j)<table(end,2)</pre>
            m(j)=4.001;
        end
    end
   %%%%2. newton method
   % m=zeros(N obs,1);
   % x0 = 0; % initial point
   % for j = 1:size(c, 1)
   %
         m(j) = newtonZ(c(j), x0);
   % end
   %%%%3. contraction mapping method
   % m=zeros(N obs,1); % initial point
   % for i = 1:size(c, 1)
          m(i) = contractionZ(m(i), c(i));
   % end
    z = eut + m;
```

```
ut searched=ut;
            searched2=repmat(searched,1,D);
            ut searched(searched2==0)=NaN;
            for d = 1:D
               %best ut_so_far
               ymax=cummax(reshape(ut searched(:,d),nalt,N cons));
               ymax=circshift(ymax,1);
               ymax=reshape(ymax, N_obs, 1);
               %best z next
                zmax=cummax(reshape(z(:,d),nalt,N_cons),'reverse');
                zmax=circshift(zmax,-1);
                zmax=reshape(zmax,N obs,1);
               %outside option for each consumer
               u0 2=ut(:,d).*outside;
               u0 3=reshape(u0 2,nalt,N cons);
               u0_4=repmat(sum(u0_3),nalt,1);
               u0_5=reshape(u0_4,N_obs,1);
                supp var = ones(size(dat,1),1);
               %selection rule: z>z next
               order=(z(:,d)-zmax).*has_searched.*searched.*(1-outside).*(1-last)
+ supp_var.*last + supp_var.*outside + supp_var.*(1-has_searched) +
supp_var.*(1-searched);
               order = (order > 0);
               %stopping rule: z>u_so_far
                search 1 = (z(:,d)-ymax).*has searched.*searched.*(1-outside) +
supp_var.*outside + supp_var.*(1-searched) + supp_var.*(1-has_searched);
               search_2 = (ymax-z(:,d)).*has_searched.*(1-searched) +
supp_var.*(1-has_searched) + supp_var.*searched;
                search_3 = (u0_5-z(:,d)).*(1-has_searched).*(1-outside) +
supp_var.*has_searched + supp_var.*outside;
                search_1 = (search_1 > 0);
                search 2 = (search 2 > 0);
                search_3 = (search_3 > 0);
               %choice rule
               u_ch2=ut(:,d).*tran;
               u_ch3=reshape(u_ch2,nalt,N_cons);
               u ch4=repmat(sum(u ch3),nalt,1);
               u_ch5=reshape(u_ch4,N_obs,1);
               choice = (u ch5-ut(:,d)).*(1-tran).*searched + supp var.*tran +
supp_var.*(1 - searched);
```

```
choice = (choice > 0);
                %1. combine all inputs
                chain_mult = order.*search_1.*search_2.*search_2.*search_3.*choice;
                %2. sum/prod at the consumer level
                prob(:,d) = accumarray(consumer, chain_mult, [N_cons 1],@prod);
            end
            %3. avg across draws (D)
            lk=mean(prob,2);
            L_ALL(:, pd, i) = lk;
        end
    end
    %Maximization
    f = \Omega(x)liklWeitz_imp(x, W_PROP, L_ALL, WEIGHT, C, consumer, Js, D, M, seed);
    %problem has no linear constraints, so set those arguments to []
    A = [];
   b = [];
    Aeq = [];
    beq = [];
    nonlcon = [];
    [be,val,exitflag,output,lambda,grad,hessian] =
fmincon(f,param,A,b,Aeq,beq,lb,ub,nonlcon,options);
    se = real(sqrt(diag(inv(hessian))));
    est = [be'; se; val; exitflag];
end
```